

Collaboration and analytics for optimal production and forecasting

Tom Fox, dynamicforecaster.com

29-Sep-2015, Aberdeen

Digital Energy Journal Forum on
Using analytics to improve production

Summary

- To maximise production we need to integrate different areas of technical expertise. We need to involve all levels in the organisation, asset equity partners and customers too.
- Integrated activity planning, rolling production forecasting and asset management must be aligned.
For credible production forecasting, we have to reach a balanced view of uncertainties.
- How can we enable such collaboration across organisations and locations?
- Integrated Operations Centres are not the whole answer.
People, Process, Technology and Organisation all need yet more attention to deliver improved workflows.
- A case-study of optimising gas-lift for many wells will show the benefits of collaboration for understanding the reservoir, production and facilities all brought together with insights from Operations.

From analytics to forecasts

Understanding the past, using analytics,
helps to forecast future production

When we have analytics,

- Who understands them?
- How will we use them?

It's a bad time to be poor at planning and forecasting

Challenges

- Low oil price
- High unit costs of mature offshore fields
- Aging platforms and pipelines infrastructure
i.e. commercial, production and maintenance

Opportunities

- Lucrative production enhancements
- Share infrastructure to reduce unit costs
- Extending field life defers cost of abandonment

Collaborative analytics & forecasting

From analytics to better forecasts

Challenges in planning and forecasting

Impact of problems

What is needed to improve forecasting?

Design of collaborative analytics & forecasting

Implementation examples

Forecasting depends on complex planning

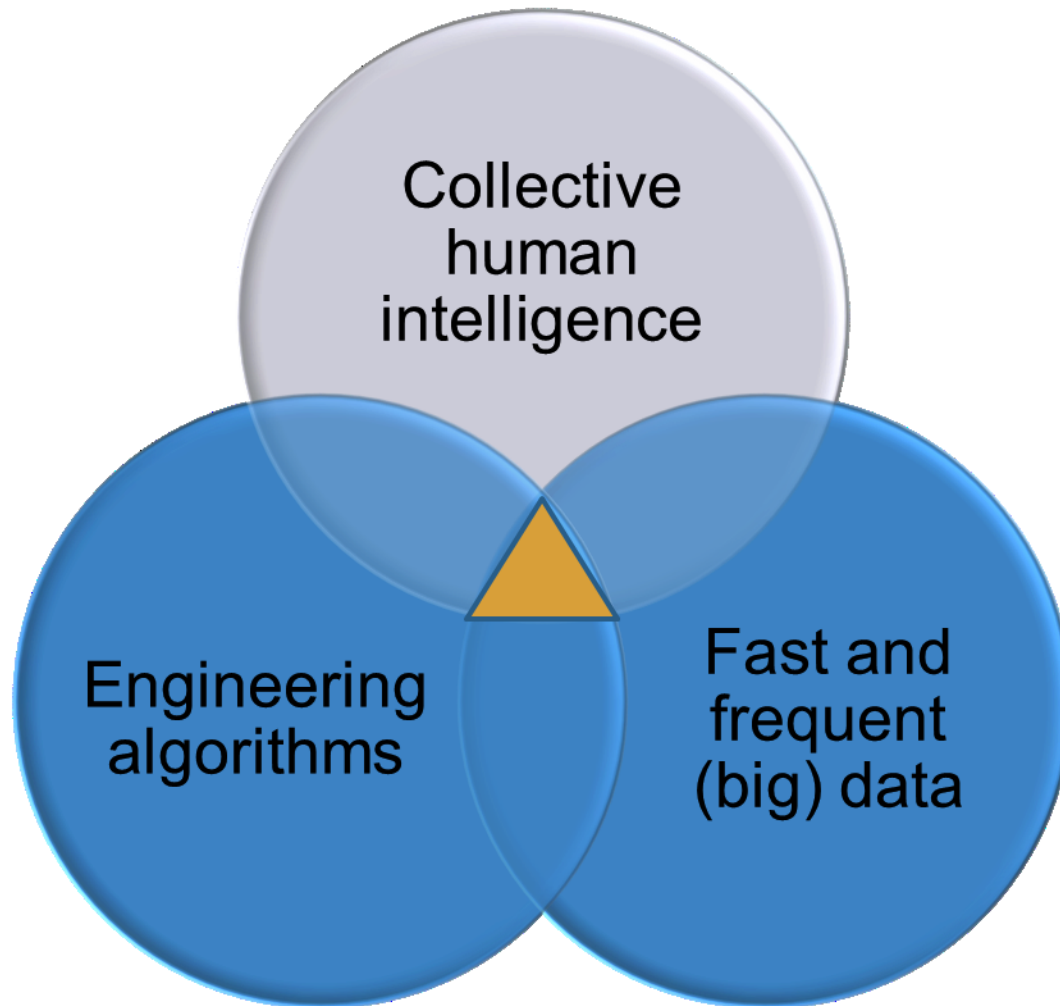
Integrated Activity Planning

- **Commercial** cash flows
- **Production** flows
- **Maintenance** logistics

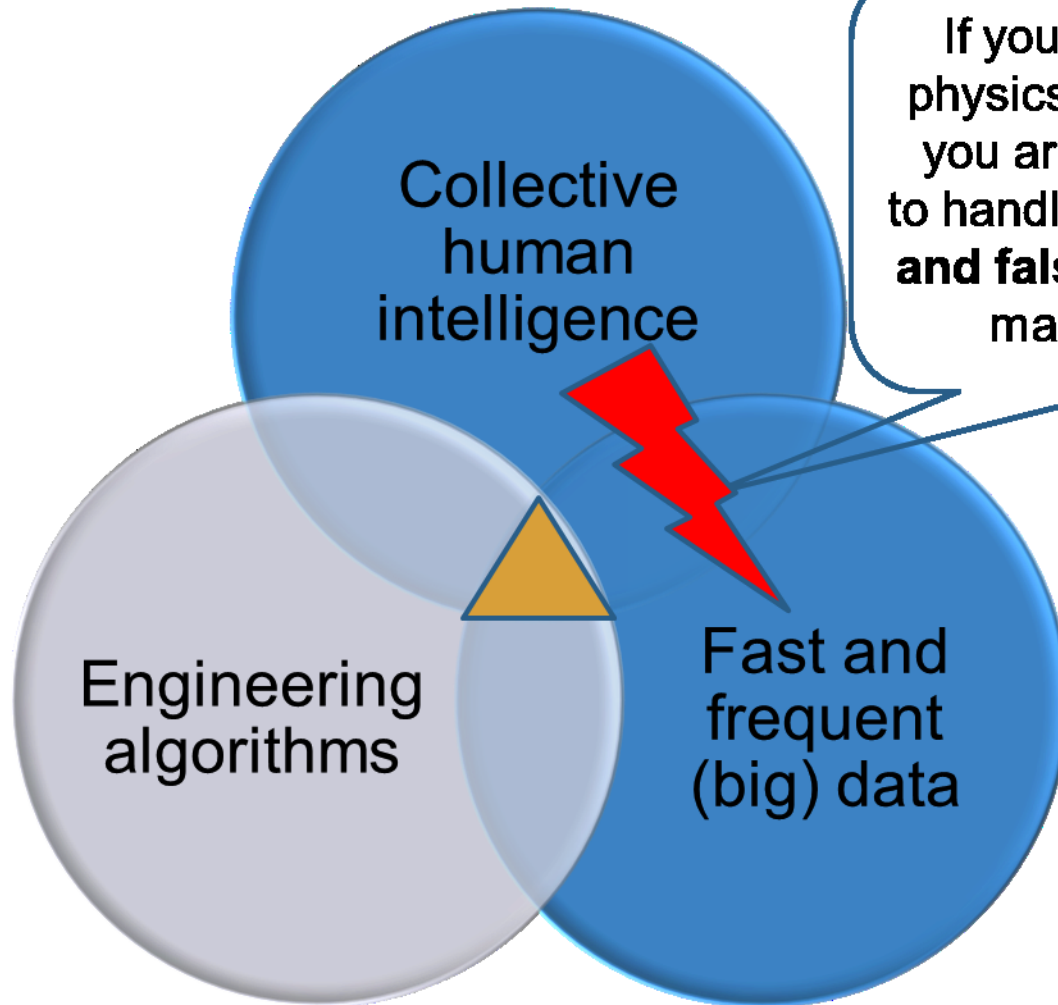
but ...

- **Complex decisions** are hard
- **Implementation** easily breaks down

A digital solution is an incomplete solution



Overloaded with misleading conclusions



Collaborative analytics & forecasting

From analytics to better forecasts

Challenges in planning and forecasting

Impact of problems

What is needed to improve forecasting?

Design of collaborative analytics & forecasting

Implementation examples

Quiz: Who can identify this ...

- ~10% of North Sea oil and gas production
- An offshore production platform; a system node
- Unusually operating in 'phase 1 mode'
- Maintenance Permit(s) to Work, e.g. PSV #504
- Hydrates in the gas compression system pipework, so stopped condensate pump B.
- Urgently started condensate pump A ...
- Disaster
- US\$1.4 billion insurance claims
- 167 lives lost in July 1988

We have not forgotten Piper A ...

Copyright image is accessible from the link below

<http://www.bbc.co.uk/news/uk-scotland-22840445>

https://en.wikipedia.org/wiki/Piper_Alpha

Missing integrated management

- ~10% of North Sea oil and gas production
- An offshore production platform; a system node
- Unusually operating in 'phase 1 mode'
- Maintenance Permit(s) to Work, e.g. PSV #504
- Hydrates in the gas compression system pipework, so stop condensate pump B.
- Urgently started condensate pump A
- Disaster
- US\$1.4 billion insurance claims
- 167 lives lost in July 1988

maintenance

production

commercial

Psychological traps impair decisions

- **Over-relying on first thoughts: the anchoring trap**
- Keeping on keeping on: the status quo trap
- Protecting earlier choices: sunk–cost trap
- **Seeing what you want to see: the confirming–evidence trap**
- Posing the wrong question: the framing trap
- Being too sure of yourself: the over-confidence trap
- Focusing on dramatic events: the recall-ability trap
- **Neglecting relevant information: the base-rate trap**
- Slanting probabilities and estimates: the prudence trap
- Seeing patterns where none exist: the out-guessing randomness trap
- Going mystical about coincidences: the surprised-by-surprises trap

Ref. Hammond, Keeney, & Raiffa, (1999). Smart Choices

Collaborative analytics & forecasting

From analytics to better forecasts

Challenges in planning and forecasting

Impact of problems

What is needed to improve forecasting?

Design of collaborative analytics & forecasting

Implementation examples

Collaboration is not a luxury

It deserves purposeful investment
of your time, energy and money

Collaboration on analysis

- Software to provide shared access to analysis
 - inputs
 - tools
 - results
- Integration of analysis
 - Multi-discipline expertise
 - Multi-role (operators, analysts, management, partners, vendors, customers).
 - Multi-location
 - Concurrent more than sequential

Communications are not a luxury

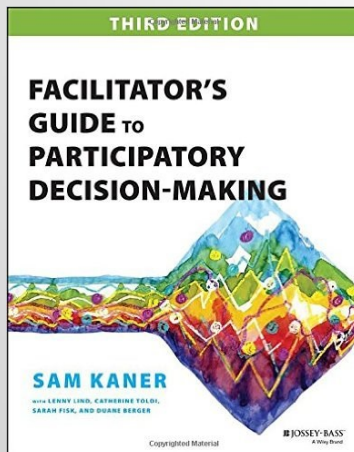
Everyone makes decisions

at all levels in your organisation.

- Why not keep them informed?
- How can they participate in complex decisions?

Gradients of agreement

is a better vocabulary than
'Yes/No'
for team decision-making



Source: Community at Work Gradients of Agreement Scale, 1996

Team Decision Making Gradients of Agreement

Enthusiastic Support

1. Fully support - *"I like it."*
2. Endorsement with minor concerns - *"Basically I like it."*

Lukewarm Support

3. Agree with reservations - *"I can live with it."*
4. Abstain - *"I have no opinion."*
5. Stand aside - *"I don't like this, but I don't want to hold up the group."*

Meager Support

6. Disagreement, but willing to go with majority - *"I want my disagreement noted, but I'll support the decision."*
7. Disagreement, with request not to be involved in implementation - *"I don't want to stop anyone else, but I don't want to be involved in implementing it."*

Strong Objection

8. Can't support the proposal

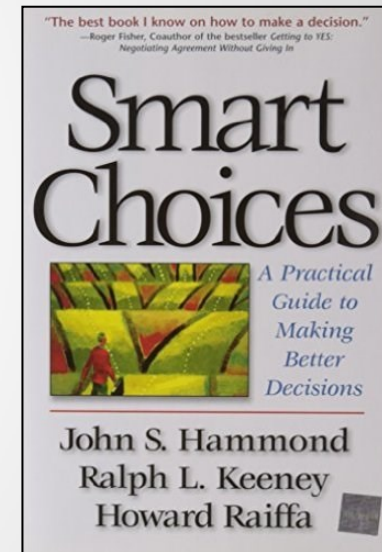
Yes

No

Smart choices: a practical guide to better decisions

ProACT

- Work on the right decision **P**roblem
- Specify your **O**bjectives
- Create imaginative **A**lternatives
- Understand the **C**onsequences
- Grapple with your **T**rade-offs
- Clarify your uncertainties
- Think hard about your risk tolerance
- Consider linked decisions
- Be aware of psychological traps



Collaborative analytics & forecasting

From analytics to better forecasts

Challenges in planning and forecasting

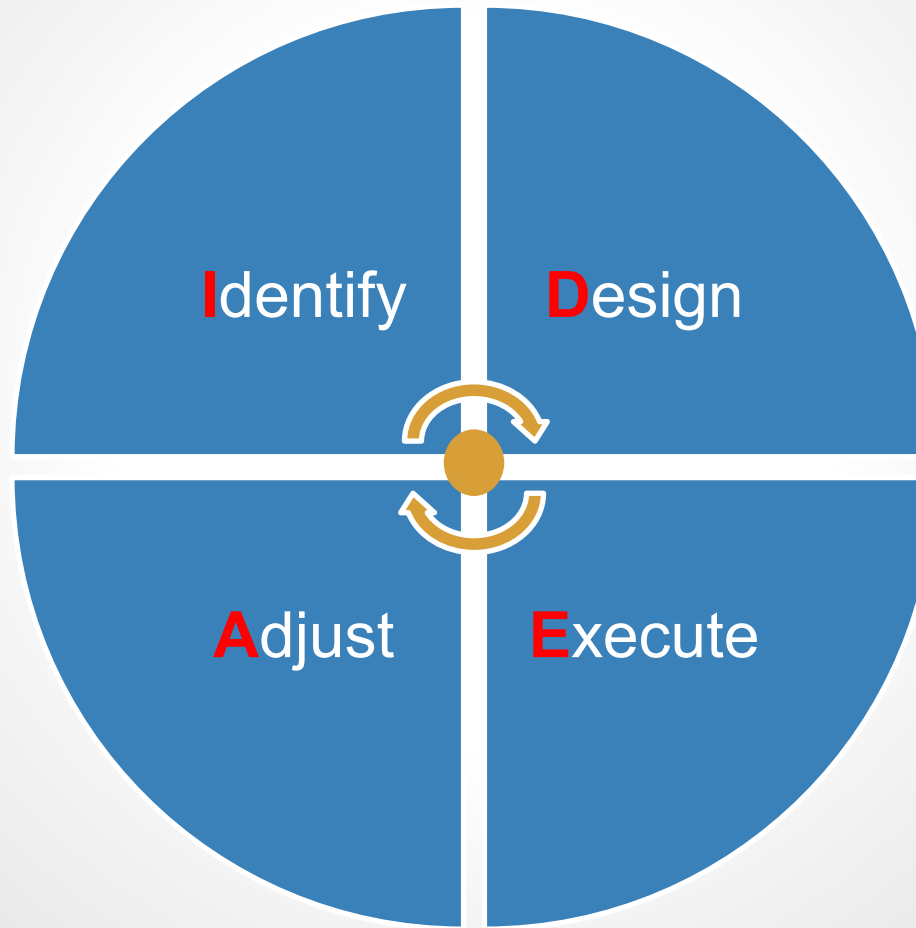
Impact of problems

What is needed to improve forecasting?

Collaborative analytics & forecasting

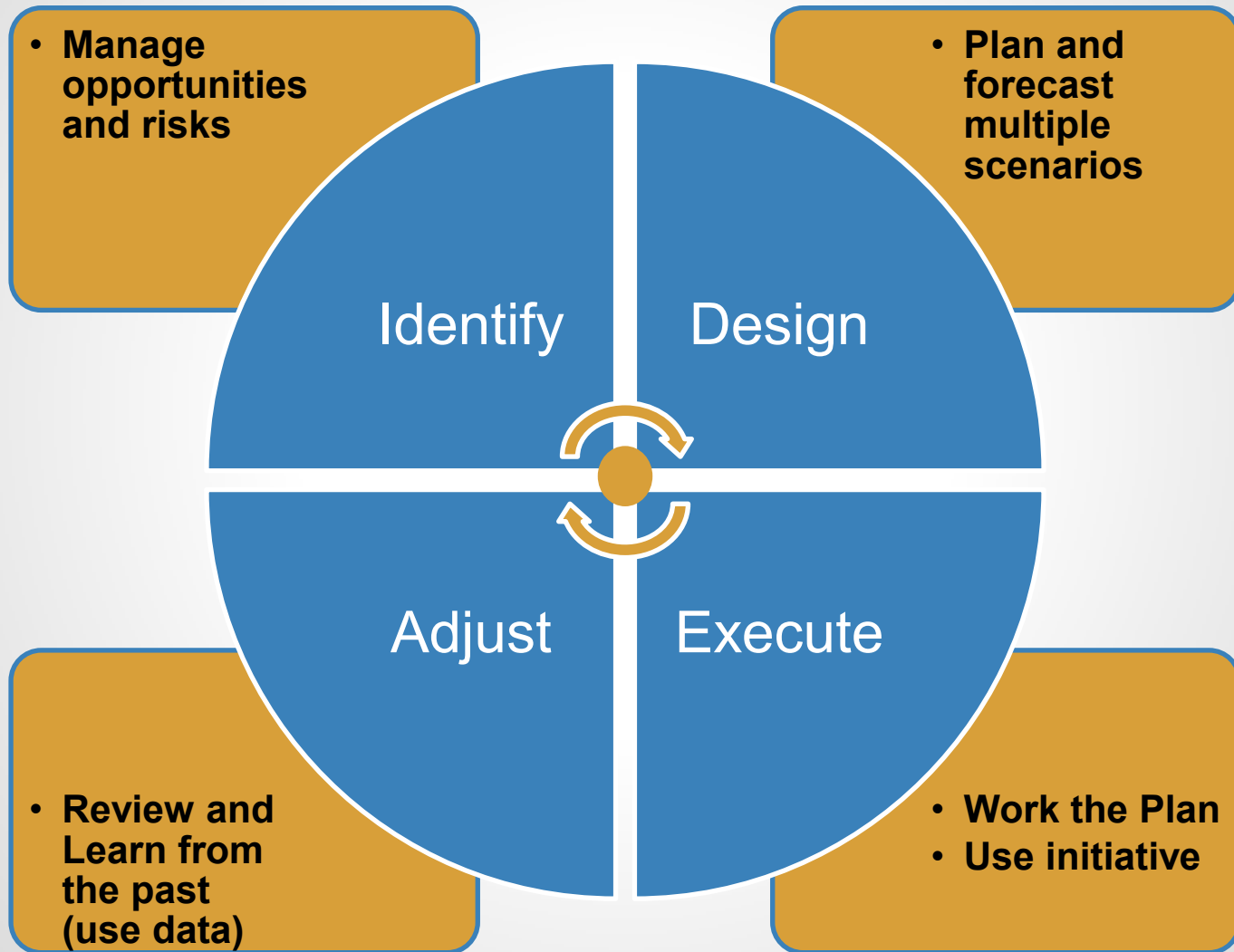
Implementation examples

'IDEA' cycle of improvement

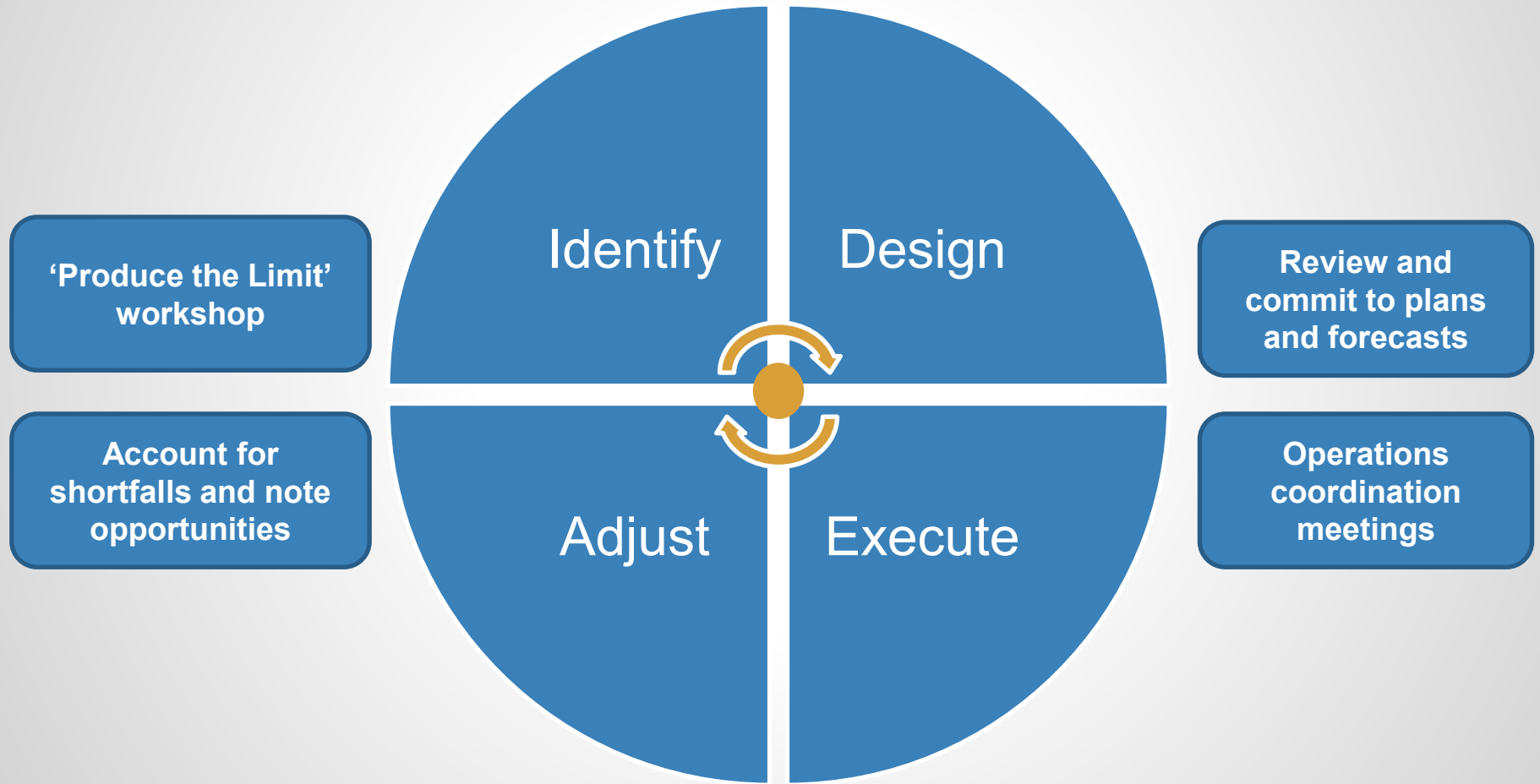


Ref. May (2007) The Elegant Solution: Toyota's Formula for Mastering Innovation

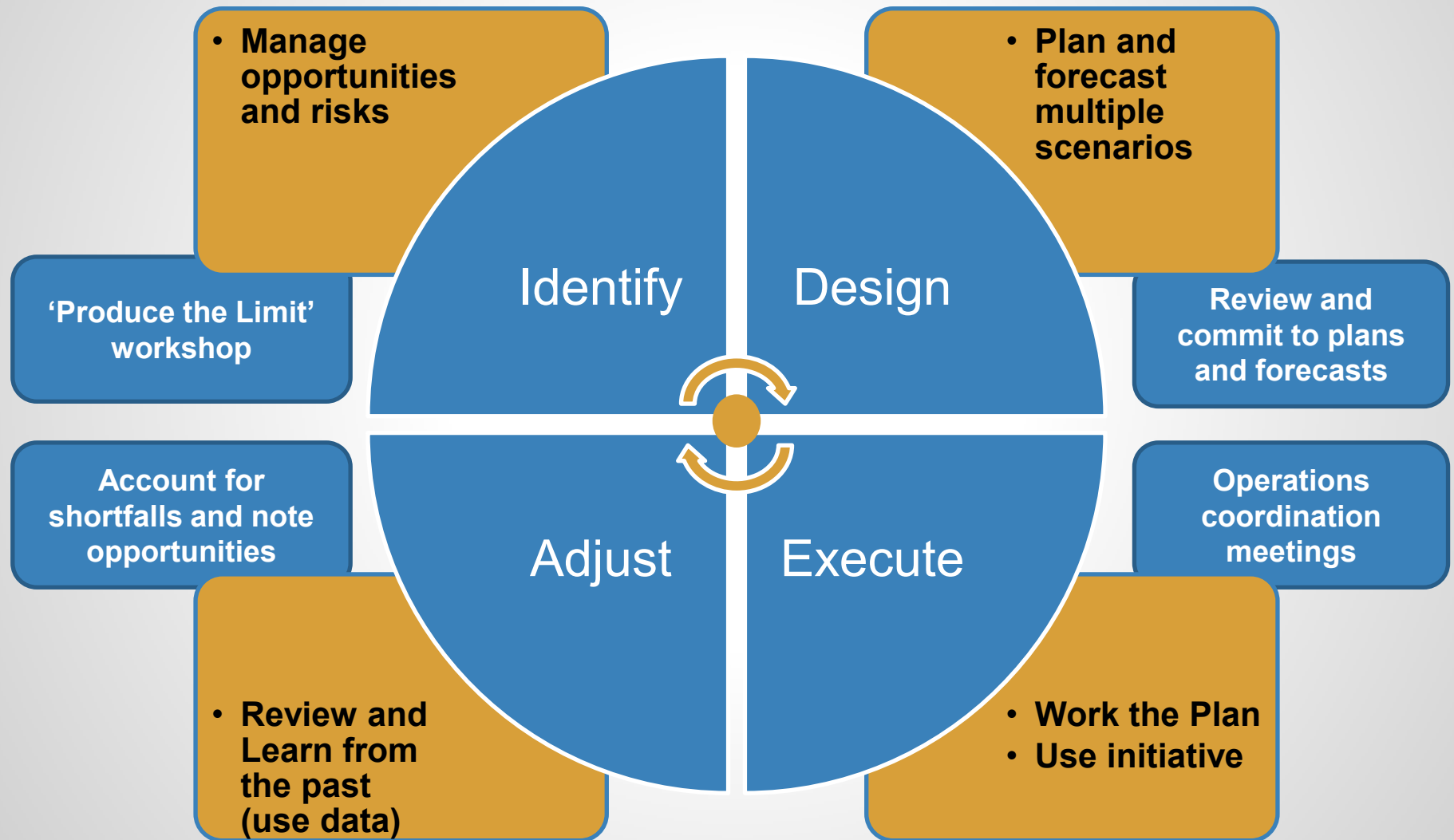
Primary processes for forecasting



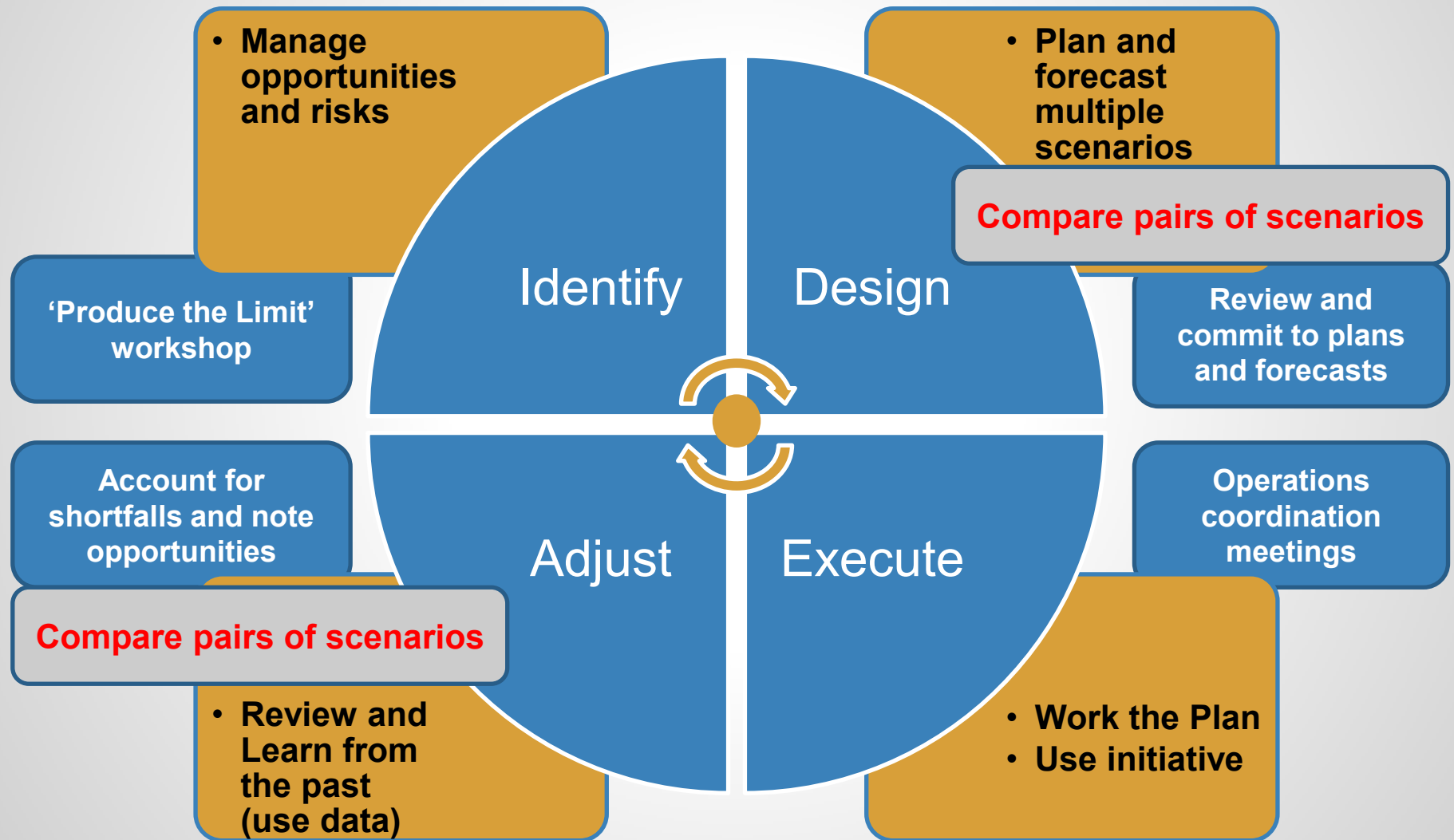
Collaborative events in planning cycle



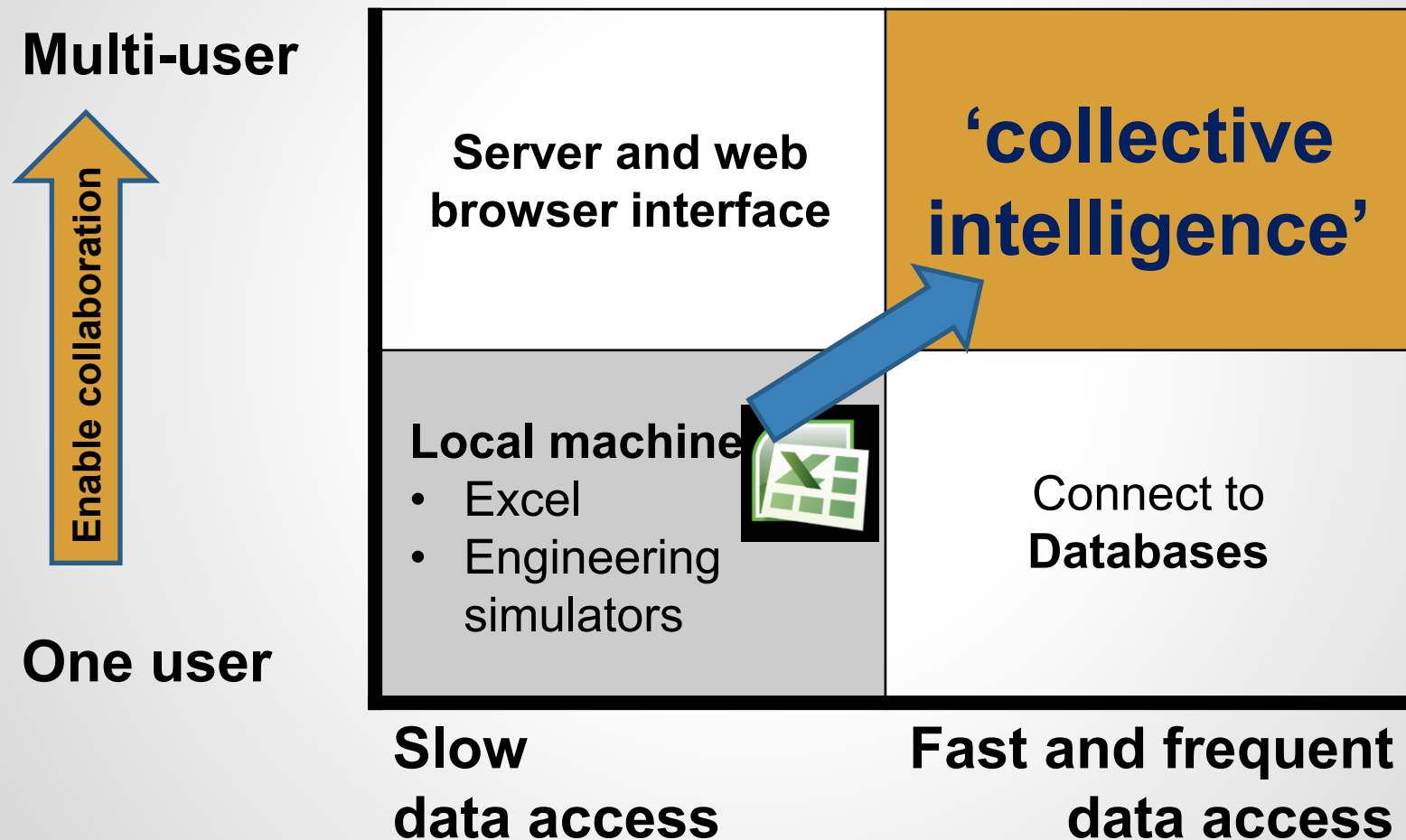
Collaborative analytics & forecasting



Collaborative analytics & forecasting

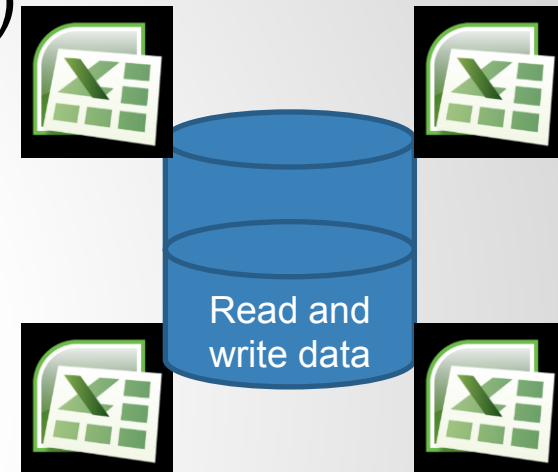


Team access to analysis tools & data



Why move beyond spreadsheets?

- Error rate is unacceptable (refs.1,2)
- Hard to enforce version control
 - VBA coding is difficult to adapt
- Lack of security for multiple users
- Poor for rolling, repetitive updates
- Risk of bad business decisions



	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	Y	AI	AJ	AT	BD
1																				
2			DateTime	TOTAL	Total_A	A_001	A_002	A_003	A_004	A_005	A_006	A_007	A_008	A_009	Total_B	Total_C	D_Total	E_Total	F_Total	
3			2013-01-01	216	45	1	2	3	4	5	6	7	8	9	27	54	90	45	45	
4			2013-01-02	216	45	1	2	3	4	5	6	7	8	9	27	54	90	45	45	
5			2013-01-03	216	45	1	2	3	4	5	6	7	8	9	27	54	90	45	45	
6			2013-01-04	216	45	1	2	3	4	5	6	7	8	9	27	54	90	45	45	
7			2013-01-05	216	45	1	2	3	4	5	6	7	8	9	27	54	90	45	45	
8			2013-01-06	216	45	1	2	3	4	5	6	7	8	9	27	54	90	45	45	
9			2013-01-07	252	45	1	2	3	4	5	6	7	8	9	27	54	126	45	81	

1. <http://panko.shidler.hawaii.edu/SSR/Myapers/whatknow.htm>
2. <http://www.eusprig.org/horror-stories.htm>

Implementation examples

DynamicForecaster, a multi-user,
web-enabled analytics solution for
collaboration on both
production analysis and forecasting

Multiple users can run analyses

DynamicForecaster

Analysis Simulation Business Networks Reservoirs

Piping and Wells Chart Groups Data Repository

Well_DCA_A4sql Forecast X

Actions

Name:	Well_DCA_A4sql Forecast
Case Group:	
Last Run:	26/09/2015 15:15:56
Date Created:	26/09/2015 15:09:58
Sensitivities Sorted:	No Sensitivities
Start Date:	01/01/2000 00:00:00
Finish Date:	19/05/2027 00:00:00
Rows Per Second:	524
Rows:	240
Data Source:	 ShaleGasSupply TG_all
Data Store:	 WellDca4ResultsSql
Status:	Run

© 2012-2015 - 1234most Ltd

CSG Well Monitor A7

localhost/DF6WA/#/DF6WA/AnalysisCase/I



DynamicForecaster

Analysis Simulation Business Networks Reservoirs

Piping and Wells Chart Groups Data Repository

CSG Well Monitor A7 X

Actions

Name:	CSG Well Monitor A7
Case Group:	
Last Run:	26/09/2015 15:16:48
Date Created:	26/09/2015 15:10:04
Sensitivities Sorted:	No Sensitivities
Start Date:	01/01/2010 00:00:00
Finish Date:	05/01/2013 00:00:00
Rows Per Second:	1504
Rows:	1097
Data Source:	 WellInputsSql
Data Store:	 WellResultsSql0
Status:	Run

© 2012-2015 - 1234most Ltd

GasLift 50: Fixed Pct

localhost/DF6WA/Home#





DynamicForecaster

Analysis Simulation Business Networks Reservoirs

Piping and Wells Chart Groups Data Repository

GasLift 50: Fixed Pct SQL-SQL X

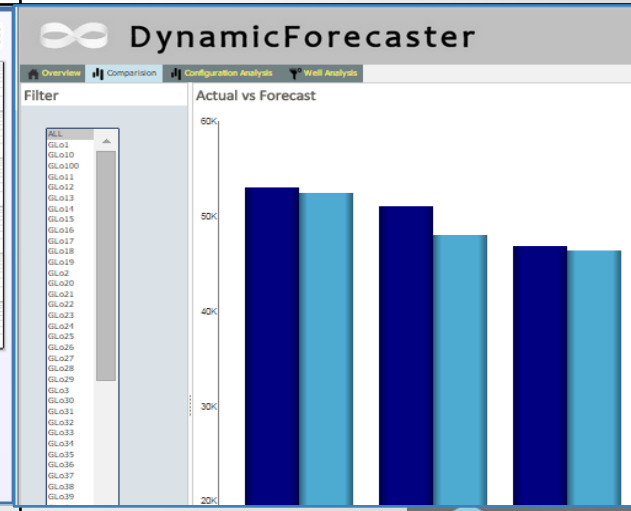
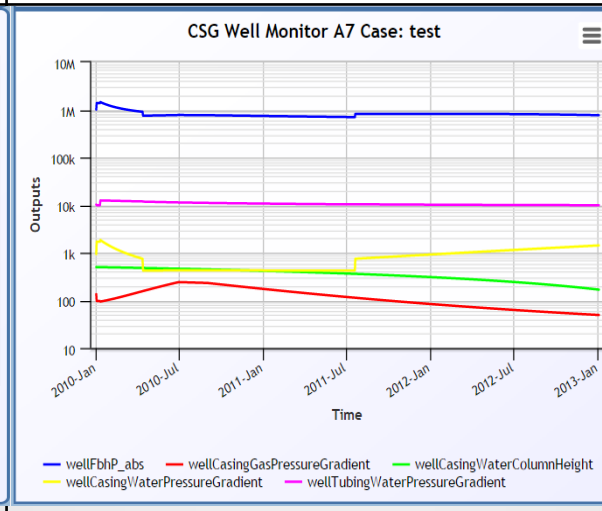
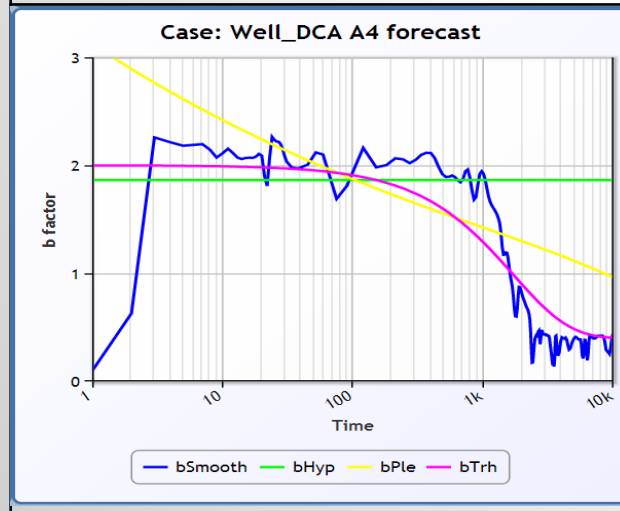
Actions

Name:	GasLift 50: Fixed Pct SQL-SQL
Case Group:	4 Demo
Last Run:	26/09/2015 15:26:17
Date Created:	26/09/2015 15:09:51
Sensitivities Sorted:	  1 Sensitivity
Start Date:	01/12/2015 00:00:00
Finish Date:	01/12/2015 23:59:59
Rows Per Second:	38
Rows:	100
Data Source:	 inputsWellsInSql
Data Store:	 outputWellsToSql
Status:	Run

© 2012-2015 - 1234most Ltd

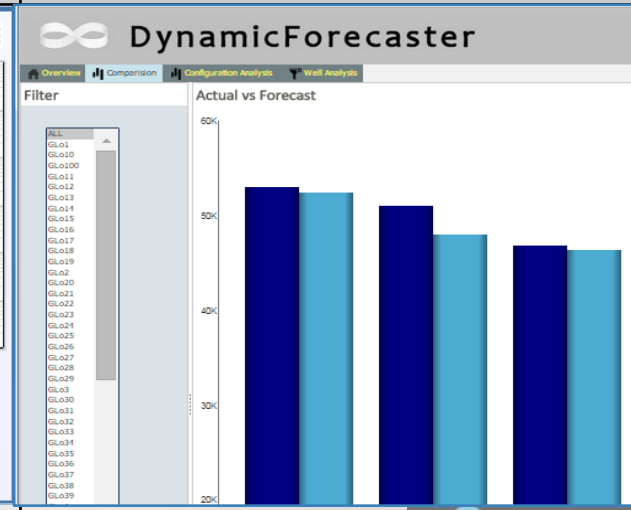
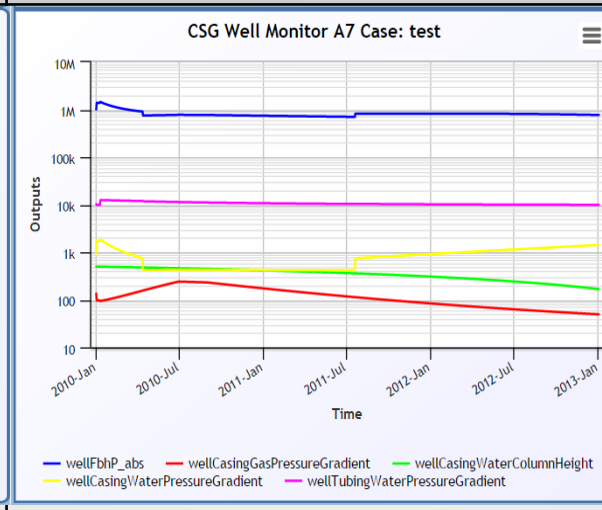
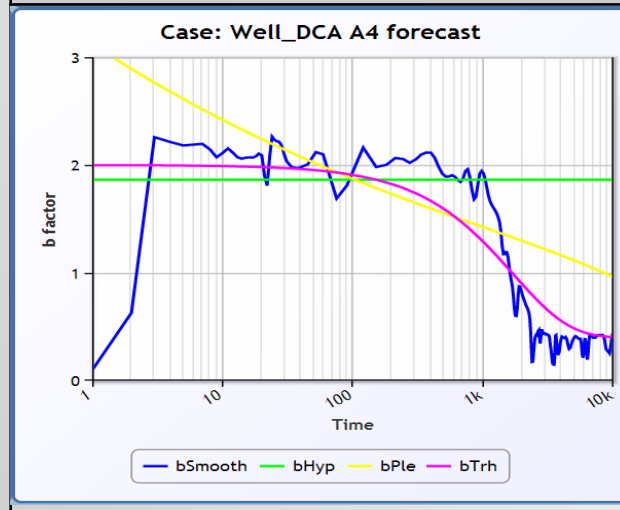
Versatile for analytics & optimized forecasts

Well forecasting	Virtual metering	Optimized scenarios
Shale Gas wells	Coal Seam Gas wells	gas-lifted Oil system
Optimal curve-fit of 4 * DCA models and then forecasts each	Computes F.B.H.Pressure and pump performance vs. expected	Computes max. oil from optimized gas-allocation for 100 wells
4 scenarios * 240 rows	1 scenario * 1100 rows	8 scenarios * 100 rows * 12 m
2 seconds (SQL-calc-SQL)	1 second (SQL-calc-SQL)	70 seconds (SQL-calc-SQL)

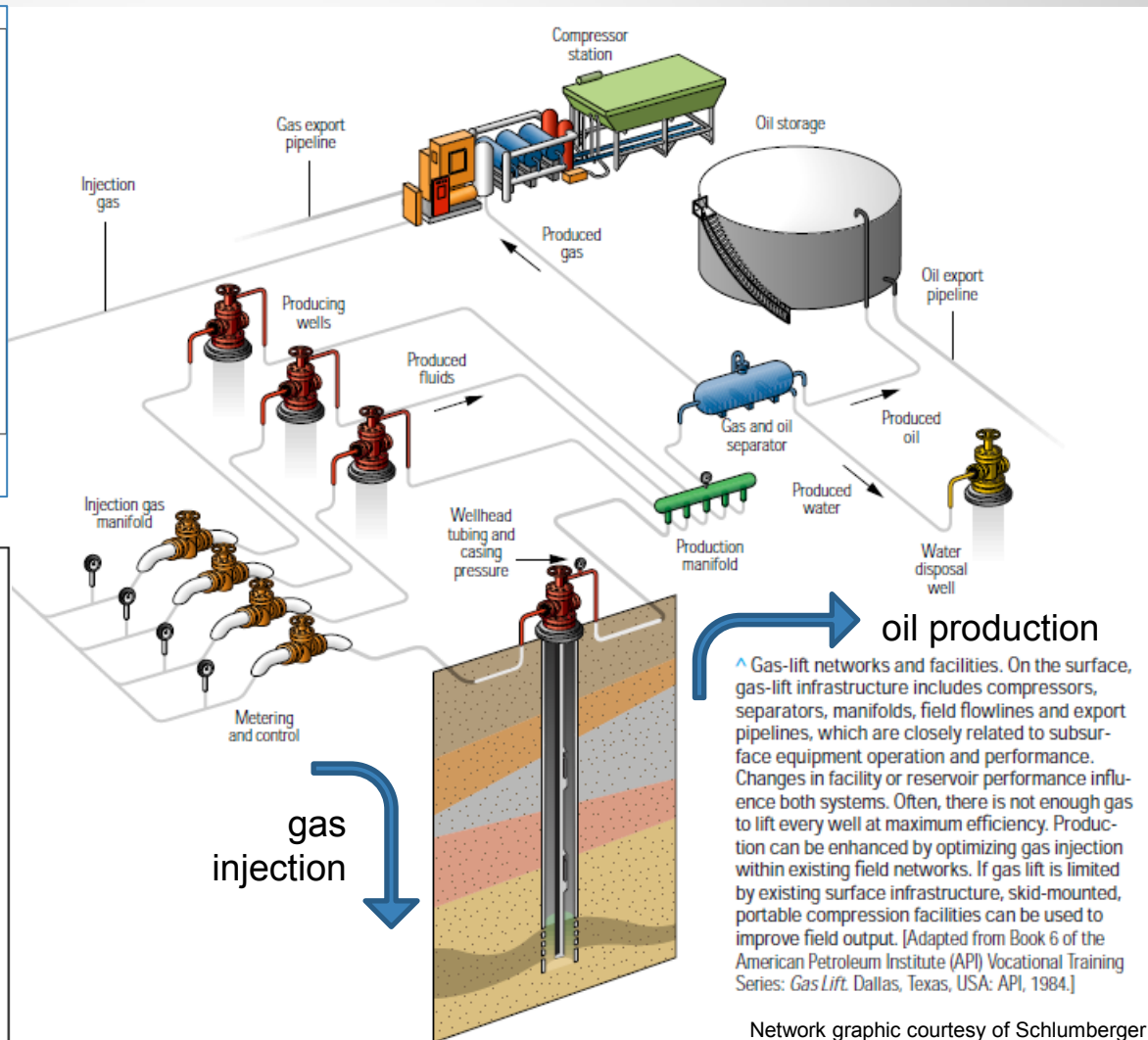
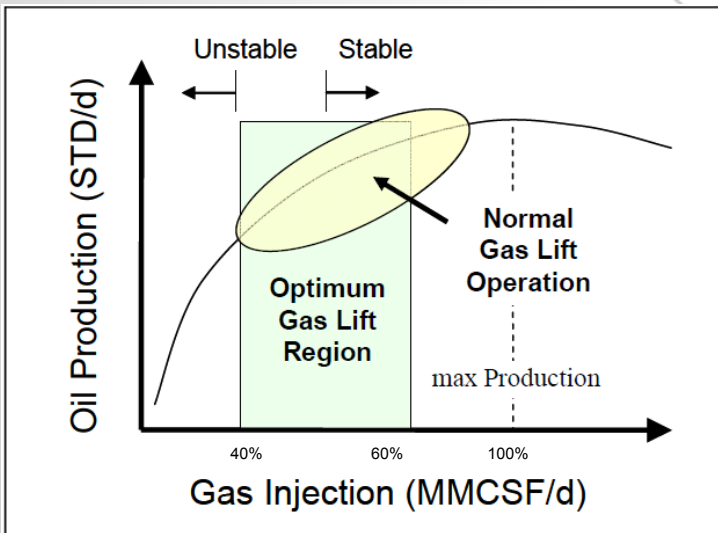
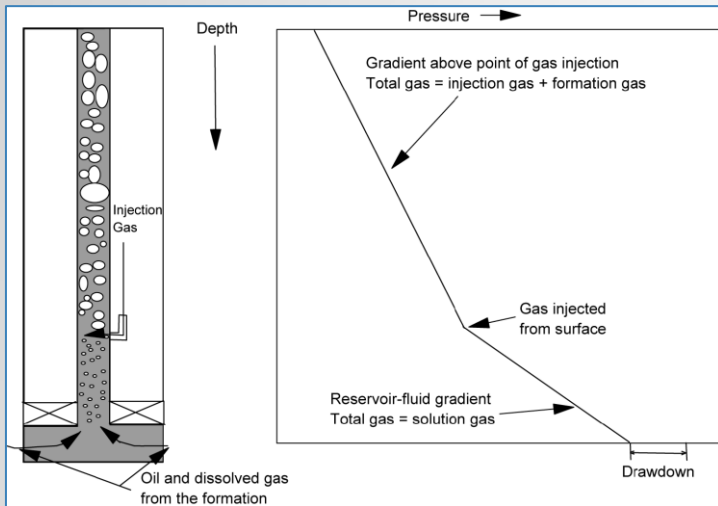


DynamicForecaster is fast with Excel I/O

Well forecasting	Virtual metering	Optimized scenarios
Shale Gas wells	Coal Seam Gas wells	gas-lifted Oil system
Optimal curve-fit of 4 * DCA models and then forecasts each	Computes F.B.H.Pressure and pump performance vs. expected	Computes max. oil from optimized gas-allocation for 100 wells
4 scenarios * 240 rows	1 scenario * 1100 rows	8 scenarios * 100 rows * 12 m
2 seconds (SQL-calc-SQL)	1 second (SQL-calc-SQL)	70 seconds (SQL-calc-SQL)
6 s 11,000 cells to Excel	2 s 30,000 cells to Excel	6 s 10,000 cells from Excel



Oil wells: maximise oil production by optimised allocation of gas-lift supply



oil production

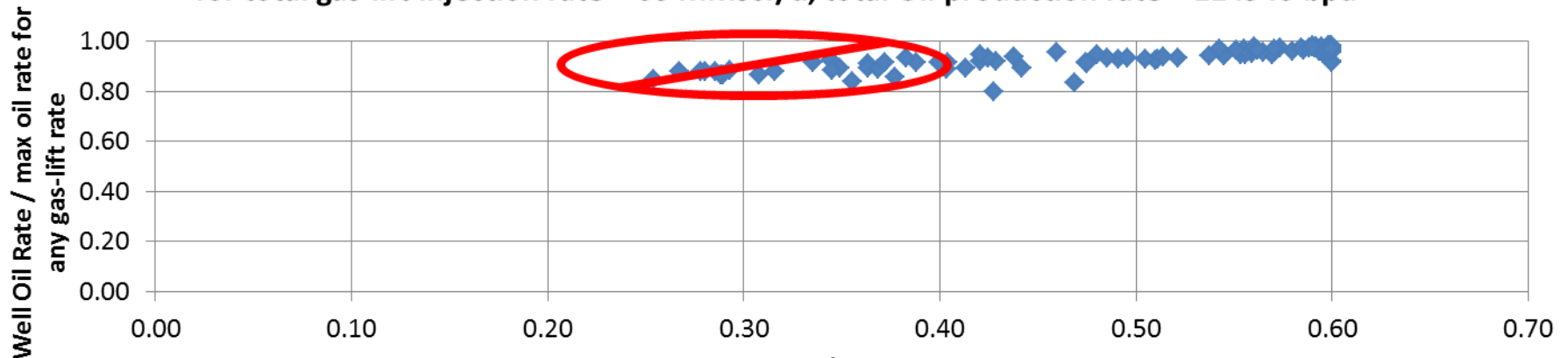
^ Gas-lift networks and facilities. On the surface, gas-lift infrastructure includes compressors, separators, manifolds, field flowlines and export pipelines, which are closely related to subsurface equipment operation and performance. Changes in facility or reservoir performance influence both systems. Often, there is not enough gas to lift every well at maximum efficiency. Production can be enhanced by optimizing gas injection within existing field networks. If gas lift is limited by existing surface infrastructure, skid-mounted, portable compression facilities can be used to improve field output. [Adapted from Book 6 of the American Petroleum Institute (API) Vocational Training Series: *Gas Lift*. Dallas, Texas, USA: API, 1984.]

Network graphic courtesy of Schlumberger

Oil wells: optimised allocation of gas-lift supply

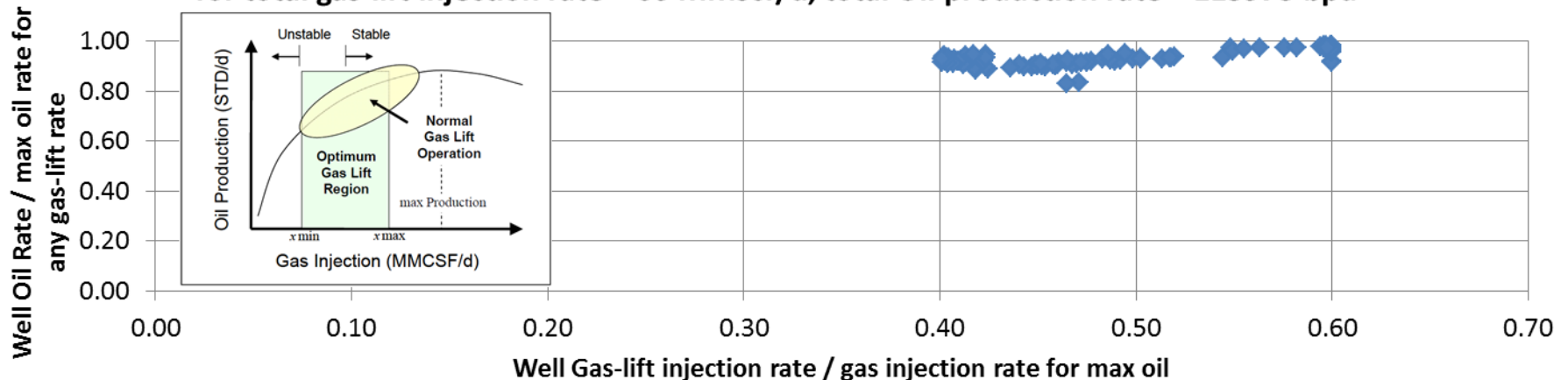
~~Optimized~~ Oil rate and Gas-Lift Rate per well

for total gas-lift injection rate = 60 MMscf/d, total Oil production rate = 114946 bpd

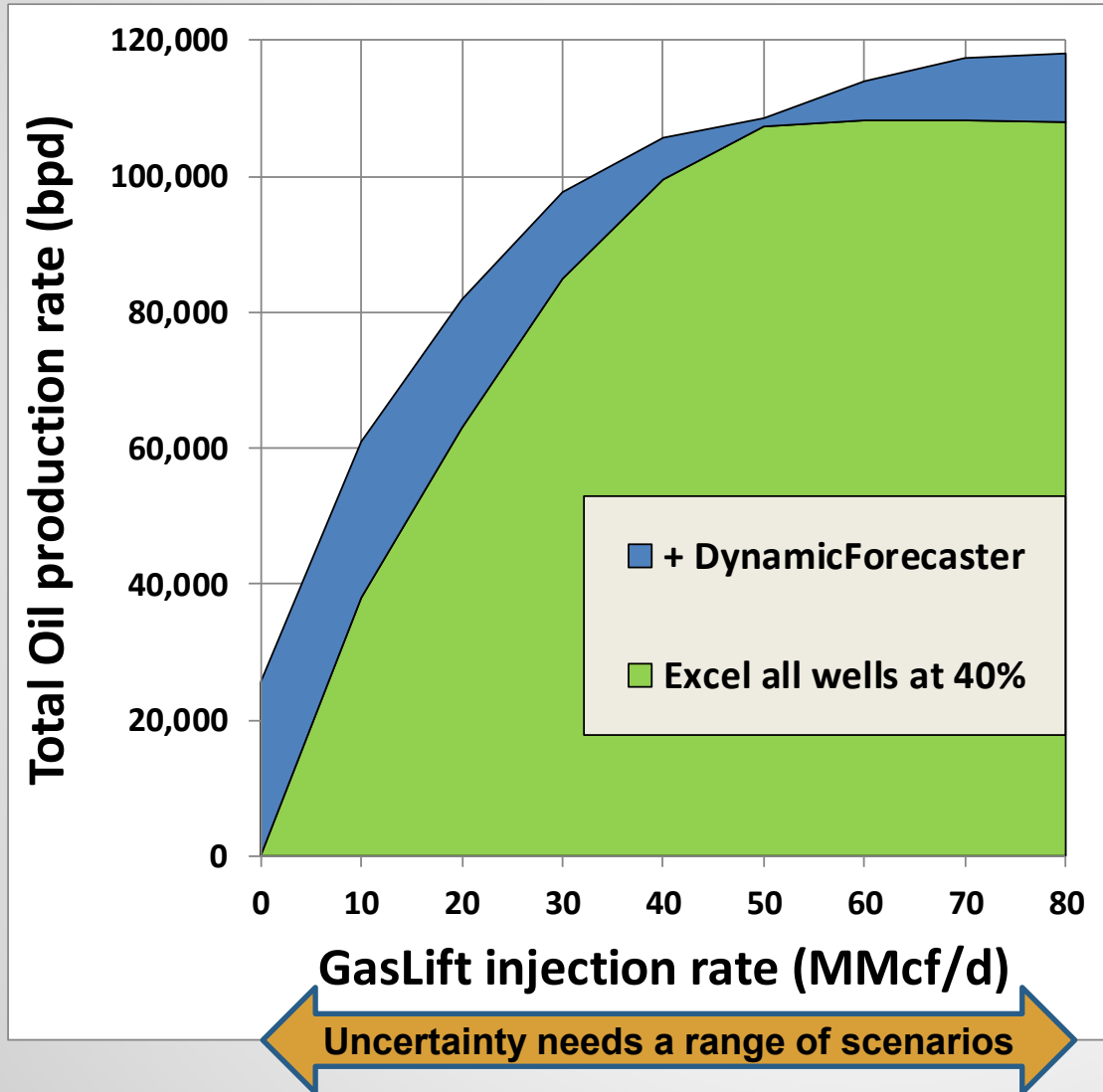


Optimized Oil rate and Gas-Lift Rate per well

for total gas-lift injection rate = 60 MMscf/d, total Oil production rate = 113978 bpd



Value-added by gas-lift optimisation



How to allow for future uncertainty in the total gas supply?

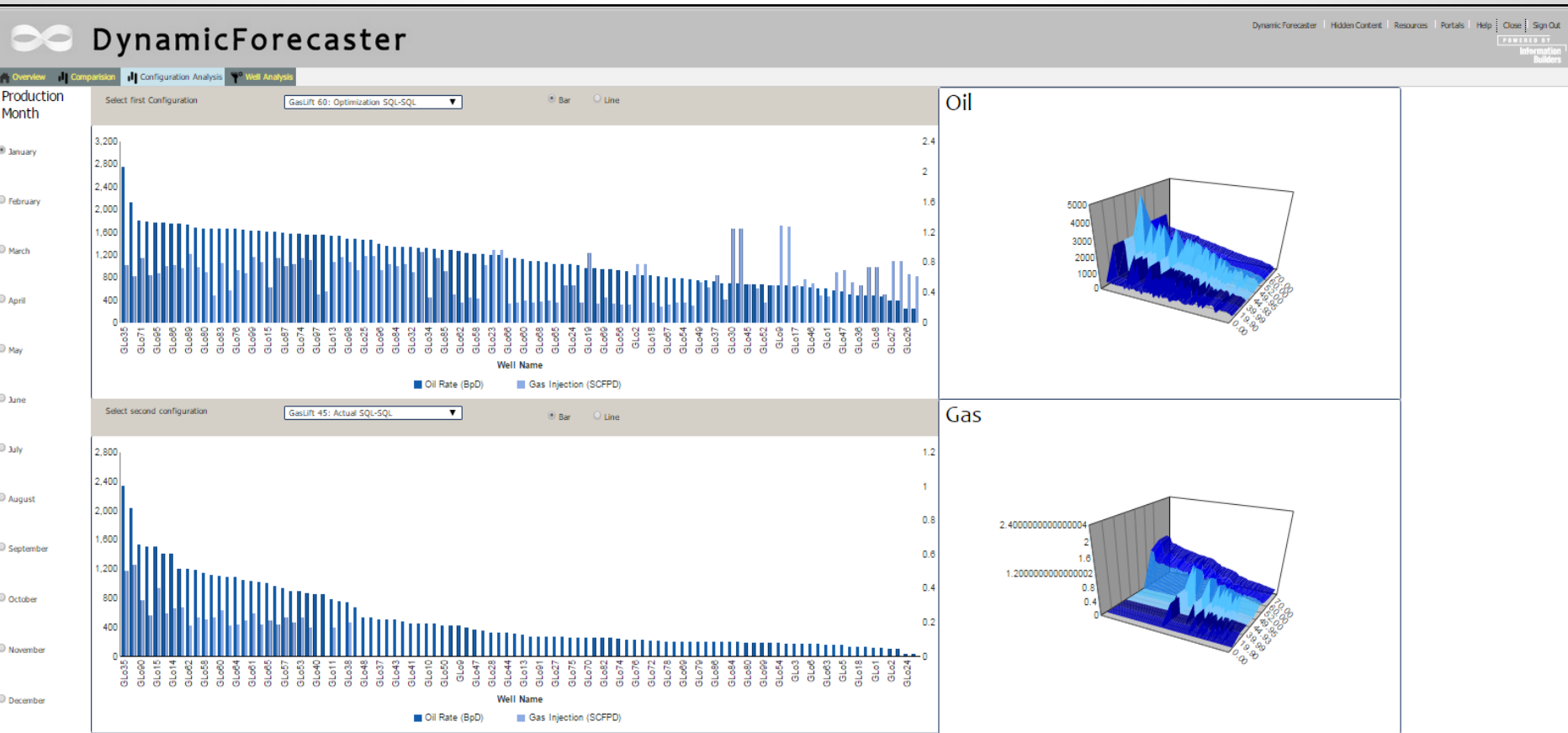
DynamicForecaster computes several *optimised* scenarios (at 0, 20, 40, 50, 60, 70 MMscf/d).

Baseline: operators are given the 40% of peak well injection gas rate for every well (perhaps from Excel)

Incremental oil should be valued at the NPV of *accelerated* production

DynamicForecaster with WebFOCUS by Information Builders

Web portal for actual production data compared with optimised gas-lift forecasts for 100 wells, 8 scenarios, monthly*12



Collaborative analytics & forecasting

Challenges in planning and forecasting

- Integrate commercial, production and maintenance

Impact of problems

- Don't have a disaster

What is needed to improve forecasting?

- Collaboration defends against psychological traps

Design of collaborative analytics & forecasting

- Processes and events for continuous improvement

Implementation examples

- Well forecasting, virtual metering, gas-lift optimisation
- High value from optimised forecasts with multiple scenarios

Know sooner, decide better, act faster

